2 1	Type BRS	1 Hits	Search Text (video near3 segment\$5) same TAPPMOG US20030165273A		USPAT; US-PGPUB; 2004/ EPO; JPO; DERWENT; 09:47 USPAT; US-PGPUB; 2004/ EPO; JPO; DERWENT; 2004/ EPO; JPO; DERWENT; 08:13
ω	IS&R	0	("US20030165273A").PN.		USPAT; US-PGPUB; EPO; JPO; DERWENT; 08:14 IBM_TDB
4	IS&R	2	("20030165273").PN.		USPAT; US-PGPUB; 2004/: EPO; JPO; DERWENT; 08:15 IBM_TDB
Уī	BRS	W	006687.ap.		USPAT; US-PGPUB; 2004/ EPO; JPO; DERWENT; 08:16 IBM_TDB
6	BRS	ω	harville-m.in.		USPAT; US-PGPUB; 2004/10/29 EPO; JPO; DERWENT; 08:16 IBM_TDB
	BRS	Н	ТАРРМОС	<u> </u>	USPAT; US-PGPUB; 2004/10/29 EPO; JPO; DERWENT; 08:19
∞	BRS	423	mixture adj3 Gaussian		USPAT; US-PGPUB; EPO; JPO; DERWENT; 08:20 IBM_TDB
9	BRS	0	"mixture of Gaussians"		USPAT; US-PGPUB; 2004/: EPO; JPO; DERWENT; 08:19 IBM_TDB
10	BRS	384	mixture adj2 Gaussian		USPAT; US-PGPUB; EPO; JPO; DERWENT; 08:21 IBM_TDB
11	BRS	199	(mixture adj2 Gaussian) with model\$4		USPAT; US-PGPUB; EPO; JPO; DERWENT; 2004/10/29 IBM_TDB
12	BRS	11	((mixture adj2 Gaussian) with model\$4) same video		USPAT; US-PGPUB; 2004/10/29 EPO; JPO; DERWENT; 08:21 IBM_TDB
13	BRS	œ	(((mixture adj2 Gaussian) with moḍel\$4) same video) and @ad<20011210	[USPAT; US-PGPUB; 2004/10/29 EPO; JPO; DERWENT; 09:28 IBM_TDB

27	26	25	24	23	22	21	20	19	18	17	16	15	14	
IS&R	BRS	BRS	BRS	BRS	BRS	BRS	BRS	BRS	BRS	BRS	BRS	BRS	BRS	Туре
12	6	9	2	21	7	151	36	37	33	34	697	0	13	Hits
(("6263088") or ("6249613") or ("6141433") or ("6075875") or ("5915044") or ("5764803")).PN.	(((video near3 segment\$5)) same (mix\$5 near3 (Gauss\$4 statistical\$2))) EPO; JPO; DERWEN and @ad<20011210 IBM_TDB	((video near3 segment\$5)) same (mix\$5 near3 (Gauss\$4 statistical\$2))	((video near3 segment\$5)) with (mix\$5 near3 (Gauss\$4 statistical\$2))	((video near3 segment\$5)) with ((Gauss\$4 statistical\$2) near3 model\$3)	((video near3 segment\$5) with background) with ((Gauss\$4 statistical\$2) near3 model\$3)	(video near3 segment\$5) with background	(((time\$1 histor\$7 temporal\$2) with (Gauss\$4 statistical\$2) with model\$3) same (video background)) and @ad<20011210	((time\$1 histor\$7 temporal\$2) with (Gauss\$4 statistical\$2) with model\$3) same (video background)	ר l fac\$3))	((time\$1 histor\$7 temporal\$2) with (Gauss\$4 statistical\$2) with model\$3) with (video surveillance intru\$4 background personnel fac\$3)	(time\$1 histor\$7 temporal\$2) with (Gauss\$4 statistical\$2) with model\$3	6188777.URPN. and model\$4	6188777.URPN.	Search Text
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2004/10/29 14:12	2004/10/29 14:10	2004/10/29 12:13	2004/10/29 12:13	2004/10/29 12:11	2004/10/29 12:10	2004/10/29 09:48	2004/10/29 12:14	2004/10/29 09:37	2004/10/29 09:37	2004/10/29 09:36	2004/10/29 09:48	2004/10/29 09:27	2004/10/29 08:39	Time Stamp
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40	39	38	37	36	35	34	33	32	31	30	29	28	
BRS	BRS	BRS	BRS	BRS	BRS	BRS	BRS	BRS	BRS	BRS	BRS	BRS	Туре
28	56	209	6	<u>н</u>	177	0	204	0	0	638	0	2	Hits
(Gaussian near5 model\$4) with (updat\$3)	(Gaussian near5 model\$4) with (updat\$3 modif\$7)	(Gaussian near5 model\$4) with (updat\$3 train\$3 modif\$7 improv\$3 revis\$3)	(hypothes\$1s model\$4) with (updat\$3 improv\$5 modif\$7) with ((positive negative) adj1 feedback)	(hypothes\$1s model\$4) with (updat\$3) with ((positive negative) adj1 feedback)	((model\$4) with (updat\$3 ((Gaussian near5 model\$4) with (updat\$3 train\$3 modif\$7 improv\$3 revis\$3) with (feedback))) with (feedback)) and (Gaussian background foreground)		(model\$4) with (updat\$3 ((Gaussian near5 model\$4) with (updat\$3 train\$3 modif\$7 improv\$3 revis\$3) with (feedback))) with (feedback)	(background near3 model\$4) with (updat\$3 train\$3 modif\$7 improv\$3 revis\$3) with (feedback)	(Gaussian near5 model\$4) with (updat\$3 train\$3 modif\$7 improv\$3 revis\$3) with (feedback)	(model\$4) with (updat\$3 train\$3 modif\$7 improv\$3 revis\$3) with (feedback)	(Gaussian adj1 model\$4) with (updat\$3 train\$3 modif\$7 improv\$3 revis\$3) with (feedback)	((("6263088") or ("6249613") or ("6141433") or ("6075875") or ("5915044") or ("5764803")).PN.) and ((background foreground) with model\$4)	Search Text
USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	DBs
2004/11/01 13:20	2004/11/01 13:20	2004/11/01 13:20	2004/11/01 13:12	2004/11/01 13:11	2004/11/01 13:09	2004/11/01 13:09	2004/11/01 13:09	2004/11/01 13:08	2004/11/01 13:14	2004/11/01 13:08	2004/11/01 13:07	2004/10/29 14:12	Time Stamp
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0		2004/11/02 16:05	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	(classifi\$7 segment\$5) with (confidence score measure\$4) with (combin\$5 merg\$3 integrat\$3)	1354 (co	BRS 1	53 E
0		2004/11/02 16:05	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	classifi\$7 with confidence with pixel) and @ad<20011210	12 (cla	BRS 1	52 E
0		2004/11/02 16:03	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	classift\$7 with confidence with pixel	16 clas	BRS	51 [
0		2004/11/02 15:54	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	(segment\$5 with confidence with pixel) and @ad<20011210		BRS 8	50
0		2004/11/02 15:53	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	segment\$5 with confidence with pixel	15 seg	BRS 1	49 [
0		2004/11/02 15:01	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	(segmentation with confidence) and @ad<20011210	64 (se	BRS (48
0		2004/11/02 15:00	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	segment\$5 near3 confidence	165 seg	BRS 1	47
0		2004/11/02 13:49	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	(segmentation with confidence) same track\$3		BRS 3	46
0		2004/11/02 14:58	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	segmentation with confidence	91 seg	BRS	5
0		2004/11/02 13:48	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	("6188777").PN.		IS&R 2	4
0		2004/11/01 15:48	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	perceptron adj1 leaming) and @ad<20011210	38 (pe	BRS	3
0		2004/11/01 15:47	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	perceptron adj1 learning	46 per	BRS	42
0		2004/11/01 15:47	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	((Gaussian near5 model\$4) with (updat\$3)) and @ad<20011210	19 ((G	BRS 1	41
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63 BRS 452 64 BRS 12 65 BRS 9	BRS BRS	BRS		62 BRS 72	61 BRS 122	60 BRS 78	59 BRS 129	58 BRS 273	57 BRS 297	56 BRS 1	55 BRS 17	54 BRS 24	Туре Ніts
@ad<20011210	ır3 (fusion combination)) with add\$6) and	((classifier) near3 (fusion combination)) with add\$6	((sensor classifier) near3 (fusion combination)) with add\$6	((classifi\$7) with (confidence score) with (combin\$5)) and @ad<20011210	(classifi\$7) with (confidence score) with (combin\$5)	((classift\$7) with (confidence score) with (combin\$5 merg\$3)) and @ad<20011210	(classifi\$7) with (confidence score) with (combin\$5 merg\$3)	(classifi\$7 segment\$5) with (confidence score) with (combin\$5 merg\$3)	(classifi\$7 segment\$5) with (confidence score) with (combin\$5 merg\$3 integrat\$3)	(classifi\$7 segment\$5) with (confidence score) with pixel with (combin\$5 merg\$3 integrat\$3)	((classift\$7 segment\$5) with (confidence score measure\$4) with pixel with (combin\$5 merg\$3 integrat\$3)) and @ad<20011210	(classifi\$7 segment\$5) with (confidence score measure\$4) with pixel with (combin\$5 merg\$3 integrat\$3)	its Search Text
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7	12	198	2	2	1-	29	40	0	0	0	0	18	Hits
(Hausdorf and template) and @ad<20011210	Hausdorf and template	Hausdorf	Hausdorf same template	Hausdorf with template	Hausdorf with template with match	((multi\$6 near3 classifier) with add\$6) and @ad<20011210	(multi\$6 near3 classifier) with add\$6	(multi\$6 near3 classifier) with (score confidence) with add\$6	(multi $\$6$ near3 classifier) with (score confidence) with (fus $\$3$ combin $\$5$) with add $\$6$	(multi\$1mod\$2 adj1 classifier) with (score confidence) with (fus\$3 combin\$5) with add\$6	(multi\$1mod\$2 adj1 classifier) with (score confidence) with (fusion combination) with add\$6	(((sensor)adj1 (fusion)) with add\$6) and @ad<20011210	Search Text
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2004/11/02 17:56	2004/11/02 17:56	2004/11/02 17:56	2004/11/02 17:56	2004/11/02 17:56	2004/11/02 17:51	2004/11/02 17:56	2004/11/02 17:12	2004/11/02 17:12	2004/11/02 17:12	2004/11/02 17:11	2004/11/02 17:10	2004/11/02 17:13	Time Stamp
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1 Tracking/detection section: Multi-level background initialization using Hidden Markov Models



Marco Cristani, Manuele Bicego, Vittorio Murino

November 2003 First ACM SIGMM international workshop on Video surveillance

Full text available: The exist 17 MB)

Additional Information: full citation, abstract, references, index terms

Most of the automated video-surveillance applications are based on the process of background modelling, aimed at discriminating motion patterns of interest at pixel, region or frame level in a nearly static scene. The issues characterizing an ordinary background modelling process are typically three: the background model representation, the initialization, and the adaptation. This paper proposes a novel initialization algorithm, able to bootstrap an integrated pixel and region-based background m ...

Keywords: Hidden Markov Model, pixel-region background initialization, video surveillance

2 Tracking: Multi-resolution background modeling of dynamic scenes using weighted match filters



Quanren Xiong, Christopher Jaynes

October 2004 Proceedings of the ACM second international workshop on Video surveillance & sensor networks

Full text available: Additional Information: full citation, abstract, references, index terms

Accurate background modeling is fundamentally important to motion-based segmentation, object tracking, and video surveillance. Models must discriminate between coherent foreground motion and periodic, random, or small pixel variations typically found in complex outdoor scenes. We introduce an adaptive match filter framework that is capable of modeling the locally changing spatial image structure. The correlation values of these filters are combined to robustly discriminate foreground regions ...

Keywords: background modeling, dynamic scenes, minimum average correlation energy filter

Modeling heterogeneous network traffic in wavelet domain Sheng Ma, Chuanyi Ji October 2001 IEEE/ACM Transactions on Networking (TON), Volume 9 Issue 5



Full text available: 📆 pdf(375,00 KB) — Additional Information: full citation, abstract, references, citings, index

Heterogeneous network traffic possesses diverse statistical properties which include complex temporal correlation and non-Gaussian distributions. A challenge to modeling heterogeneous traffic is to develop a traffic model which can accurately characterize these statistical properties, which is computationally efficient, and which is feasible for analysis. This work develops wavelet traffic models for tackling these issues. In specific, we model the wavelet coefficients rather than the original t ...

Keywords: Long-range dependence, network traffic modeling, self-similar traffic, wavelets

Poster: Gaussian Mixture Models for on-line signature verification

Jonas Richiardi, Andrzej Drygajlo

November 2003 Proceedings of the 2003 ACM SIGMM workshop on Biometrics methods and applications

Additional Information: full citation, abstract, references, index terms Full text available: (378,21 KB)

This paper introduces and motivates the use of Gaussian Mixture Models (GMMs) for on-line signature verification. The individual Gaussian components are shown to represent some local, signer-dependent features that characterise spatial and temporal aspects of a signature, and are effective for modelling its specificity. The focus of this work is on automated order selection for signature models, based on the Minimum Description Length (MDL) principle. A complete experimental evaluation of the Ga ...

Keywords: Gaussian Mixture Models, Hidden Markov Models, biometrics, model order, online signature, signature verification

5 Image Models

Narendra Ahuja, B. J. Schachter

December 1981 ACM Computing Surveys (CSUR), Volume 13 Issue 4

Full text available: gdf(2.99 MB)

Additional Information: full citation, references, citings, index terms

6 Special issue on independent components analysis: A generative model for separating illumination and reflectance from images

Inna Stainvas, David Lowe

December 2003 The Journal of Machine Learning Research, Volume 4

Full text available: 📆 odf 764.42 KB) — Additional Information: full odation, abstract, index terms

It is well known that even slight changes in nonuniform illumination lead to a large image variability and are crucial for many visual tasks. This paper presents a new ICA related probabilistic model where the number of sources exceeds the number of sensors to perform an image segmentation and illumination removal, simultaneously. We model illumination and reflectance in log space by a generalized autoregressive process and Hidden Gaussian Markov random field, respectively. The model ability to d ...

7 Recognition: Real-time and accurate segmentation of moving objects in dynamic scene Tao Yang, Stan Z. Li, Quan Pan, Jing Li

October 2004 Proceedings of the ACM second international workshop on Video surveillance & sensor networks

Full text available: Red pdf(644.18 KB) Additional Information: full citation, abstract, references, index terms

Fast and accurate segmentation of moving objects in video sequences is a basic task in many computer vision and video analysis applications. It has a critical impact on the performance of object tracking and classification and activity analysis. This paper presents effective methods for solving this problem. Firstly, a fast and efficient algorithm is presented for background update to handle various sources of scene changes, including ghosts, left objects, camera shaking, and abrupt illuminat ...

Keywords: background modeling, foreground segmentation, video processing, video surveillance

System section: 3D video surveillance with Augmented Virtual Environments Ismail Oner Sebe, Jinhui Hu, Suya You, Ulrich Neumann November 2003 First ACM SIGMM international workshop on Video surveillance



Full text available: pdf(583.25 KB) Additional Information: full cliation, abstract, references, index terms

Recent advances in sensing and computing technologies have inspired a new generation of data analysis and visualization systems for video surveillance applications. We present a novel visualization system for video surveillance based on an Augmented Virtual Environment (AVE) that fuses dynamic imagery with 3D models in a real-time display to help observers comprehend multiple streams of temporal data and imagery from arbitrary views of the scene. This paper focuses on our recent technical extens ...

Keywords: augmented reality, object detection and tracking, video surveillance

9 On the propagation of long-range dependence in the Internet



A. Veres, Kenesi S. Molnár, G. Vattay

August 2000 ACM SIGCOMM Computer Communication Review , Proceedings of the conference on Applications, Technologies, Architectures, and Protocols for Computer Communication, Volume 30 Issue 4

Full text available: The paint (1 34 MB)

Additional Information: full citation, abstract, references, citings, index terms

This paper analyzes how TCP congestion control can propagate self-similarity between distant areas of the Internet. This property of TCP is due to its congestion control algorithm, which adapts to self-similar fluctuations on several timescales. The mechanisms and limitations of this propagation are investigated, and it is demonstrated that if a TCP connection shares a bottleneck link with a self-similar background traffic flow, it propagates the correlation structure of the background traf ...

Keywords: TCP adaptivity, TCP congestion control, long-range dependence, self-similarity

10 Jeremiah: the face of computer vision

Richard Bowden, Pakorn Kaewtrakulpong, Martin Lewin June 2002 Proceedings of the 2nd international symposium on Smart graphics

Additional Information: full citation, abstract, references, citings, index

Full text available: pkif(4 69 MB)

terms

This paper presents a humanoid computer interface (Jeremiah) that is capable of extracting moving objects from a video stream and responding by directing the gaze of an animated head toward it. It further responds through change of expression reflecting the emotional state of the system as a response to stimuli. As such, the system exhibits similar behavior to a child. The system was originally designed as a robust visual tracking system capable of performing accurately and consistently within a ...